



IDENTIFICATION OF SELECTIVE FORWARDING ATTACKS USING CHANNEL AWARE AND ADAPTIVE SCHEME IN WSN

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Abstract

As a promising event monitoring and data gathering technique, wireless sensor network (WSN) has been widely applied to both military and civilian applications. Many WSNs are deployed in unattended and even hostile environments to perform mission-critical tasks, such as battle field reconnaissance and homeland security monitoring. So due to the lack of physical protection, sensor nodes are easily compromised by adversaries. Wireless sensor networks (WSNs) are vulnerable to selective forwarding attacks that can maliciously drop a subset of forwarding packets to degrade network performance. Meanwhile, due to the unstable wireless channel in WSNs, the packet loss rate during the communication of sensor nodes may be high and vary from time to time. It poses a great challenge to distinguish the malicious drop and normal packet loss. A Channel-aware Reputation System with adaptive detection threshold (CRS-A) can detect selective forwarding attacks in WSNs. The CRS-A evaluates the data forwarding behaviors of sensor nodes, according to the deviation of the monitored packet loss and the estimated normal loss. An attack-tolerant data forwarding scheme is developed to collaborate with CRS-A for stimulating the forwarding cooperation of compromised nodes and improving the data delivery ratio of the network. But these solutions for wireless networks may not always be sufficient. Because some malicious nodes pretend to be intermediate nodes of a route to some given destinations, drop any packet that



subsequently goes through it, is one of the major types of attack. In this paper, in addition to CRS-A, uses Ad-hoc on demand Distance Vector (AODV) routing that propose a co-operative method to detect malicious node effectively.

Keywords: WSN, AODV, CRS-A, malicious node, reputation value, co operative approach

1. Introduction

As a promising event monitoring and data gathering technique, wireless sensor network (WSN) [1] has been widely applied to both military and civilian applications. Many WSNs are deployed in unattended and even hostile environments to perform mission-critical tasks, such as battle field reconnaissance and homeland security monitoring. However, due to the lack of physical protection, sensor nodes are easily compromised by adversaries, making WSN vulnerable to various security threats. One of the most severe threats is selective forwarding attack, where the compromised nodes can maliciously drop a subset of forwarding packets to deteriorate the data delivery ratio of the network. It also has significantly negative impacts to data integrity, especially for data-sensitive applications, e.g., health-care and industry monitoring. On the other hand, since WSNs are generally deployed in open areas (e.g., primeval forest), the unstable wireless channel and medium access collision can cause remarkable normal packet losses. The selective forwarding attacks are concealed by the normal packet losses, complicating the attack detection. Therefore, it is challenging to detect the selective forwarding attacks and improve the network performance.

In this paper, proposed Channel-aware Reputation System with adaptive detection threshold (CRS-A) [1] [2] [3] [4] to detect selective forwarding attacks in WSNs with the detection of malicious node. Specifically, we divide the network lifetime to a sequence of evaluation periods. During each evaluation period, sensor nodes estimate the normal packet loss rates between themselves and their neighboring nodes, and adopt the estimated packet loss rates to evaluate the forwarding behaviors of its downstream



neighbors along the data forwarding path [4] [5]. The sensor nodes misbehaving in data forwarding are punished with reduced reputation values by CRS-A. Once the reputation value of a sensor node is below an alarm value, it would be identified as a compromised node by CRS-A. In the malicious node detection phase, each node transmits data to a next node, stores a copy of the data in its buffer and overhears whether the next node transmits the data. If the node overhears data transmission of the next node within a predetermined length of time, the node considers that the data was properly transmitted and deletes the copy of the data from the buffer. If not so, the node increases a failure tally for the next node. If the failure tally is greater than a threshold, the node determines that the next node intentionally dropped the data and reports this fact to all nodes over the network [6] [7]. The mechanism is cooperative because nodes in the protocol work cooperatively together so that they can analyze, detect malicious nodes in a reliable manner.

2. Proposed System model and Design goals

We consider a WSN consisting of a set of randomly distributed sensor nodes, denoted by N , and a sink node to monitor an open area. Each sensor node periodically senses the interested information from the surroundings, and transmits the sensed data to the sink via multi-hop routing among sensor nodes. Sensor nodes communicate with their neighboring nodes based on the IEEE 802.11 DCF. The monitored area has an unstable radio environment, making the packet loss rates during the communications of sensor nodes significantly increased and vary from time to time. Since sensor nodes are deployed in open area and lack adequate physical protection, they may be compromised by adversaries through physical capture or software vulnerabilities to misbehave in data forwarding. We use PM to denote the compromising probability of sensor node, which is defined as the probability that a sensor node is compromised by the adversary. Meanwhile, we assume that sensor nodes can monitor the data forwarding traffic of their neighboring nodes by neighbor monitoring with Watchdog or acknowledgment-based approaches. It means that a sensor node can obtain that how many data packets are forwarded by its forwarding sensor nodes. Existing works provide a comprehensive study



on monitoring forwarding traffic of sensor nodes, which is not the focus of this paper. Since the unstable radio environment causes fluctuated packet loss rates between the neighboring nodes, it is challenging to distinguish the monitored forwarding behavior is normal or not [8] [9] [10].

Compromised sensor nodes can launch selective forwarding attacks to degrade the performance of the network. Specifically, when a compromised sensor node receives a data packet, it maliciously drops it with a probability, referred to as attack probability. Since the adversary can control the attack probabilities of compromised nodes, it is difficult to distinguish if the packet losses are caused by fluctuated channel condition or malicious drops, especially for the nodes with low attack probabilities.

Furthermore, several neighboring compromised sensor nodes can collaborate with each other to launch promotion/demotion attacks to achieve benefits. For example, if N_a and N_b are two neighboring compromised sensor nodes and data traffic is from N_a to N_b , N_a may provide a partial evaluation for N_b 's forwarding behaviors. Besides, N_a can announce N_b as a normal node to its other neighboring nodes, in spite of N_b misbehaving in the data forwarding. However, we do not consider the special case where N_a is totally honest in data forwarding to cover for N_b 's misbehaviors to achieve benefits. This case can be effectively addressed by the hop-by-hop acknowledgment or two directional neighbor monitoring techniques.

3. Proposed Channel Aware Reputation System with Adaptive Detection Threshold for Detecting selective forwarding attacks

In CRS-A, each sensor node maintains a reputation table to evaluate the long-term forwarding behaviors of its neighboring nodes. The essence of CRS-A is to dynamically update the reputation table based on the forwarding behavior evaluation for the neighboring nodes, by taking the normal packet loss rate into consideration. However, as the unstable radio environment make the quality of wireless channel vary with time, normal packet loss may be different over a long time period. Therefore, we divide the



whole network lifetime into a sequence of evaluation periods $T = \{T_1, \dots, T_t, \dots\}$. In each evaluation period T_t , the channel condition of each data transmission link is assumed to be stable. Meanwhile, for each T_t , we introduce a channel estimation stage at the beginning of T_t , and a reputation update stage at the end of T_t .

During the channel estimation stage, sensor nodes estimate the normal packet loss rates of the communication links with their neighboring nodes, and use them to evaluate the forwarding behaviors of neighboring nodes. The reputation update in CRS-A consists of three procedures: reputation evaluation, propagation and integration. Reputation Evaluation is to evaluate short-term reputation scores for the forwarding behaviors of sensor nodes, based on the deviation of estimated normal packet loss rate and monitored actual packet loss rate. With Reputation Propagation, the evaluated short-term reputation scores can be propagated within the neighboring nodes to achieve a more comprehensive evaluation. Finally, by Reputation Integration, sensor nodes integrate the reputation scores evaluated by them and the propagated reputation scores from their neighboring nodes to update the reputation table.

3.1 Normal Packet Loss Estimation

According to the network model, normal packet loss is mainly caused by the poor and unstable wireless channel and MAC layer collisions. The poor and unstable radio link quality is the primary reason for the time-varied packet losses. It is formulated as a two-state Markov model, and the packet loss rate is determined as an average value over a long-term period. However, adopting an average value to represent a time-varied value may mislead the evaluation for forwarding behaviors. Furthermore, dynamic environments make the link quality varied in different locations. Therefore, the packet loss estimation should be performed in each evaluation period by each sensor node. In CRS-A, the link quality estimation for each pair of neighboring nodes is based on the Received Signal Strength Indicator (RSSI) and Signal-to-Noise Ratio (SNR), under the symmetric channel assumption. For each T_t , the packet loss rate caused by poor link



quality, denoted by $p_{ij}^1(t)$, can be estimated by RSSI and SNR for the transmission link from N_i to N_j .

As data transmission between two neighboring nodes is based on the IEEE 802.11, MAC layer collisions may increase the normal packet loss rate. Since sensor nodes are static in our network, it means each sensor node has a fixed number of neighboring nodes. Then, we can use the analytical results in to estimate the packet loss caused by medium access collisions without the impact of hidden terminals. Let n be the number of nodes contending for channel access at N_j and p_t as the probability that a node transmits data in time slot. When MAC channel is at steady state, the probabilities for observing an idle, successful, and colliding slot, denoted as p_i , p_s , and p_c , respectively, are

$$p_i = (1-p_t)^n$$

$$p_s = n \cdot p_t (1-p_t)^{n-1}$$

$$p_c = 1 - p_i - p_s$$

And the channel busy ratio R_b can be calculated as

$$C_b = 1 - (p_i \cdot t_d) / (p_i \cdot \sigma + p_s \cdot t_s + p_c \cdot t_c)$$

where t_d , t_s and t_c denote the idle slot length, the duration of a successful transmission, and the duration of a collision, respectively.

3.2 Reputation Evaluation

In CRS-A, sensor nodes monitor their neighbors to evaluate reputation scores for their forwarding behaviors during each evaluation period. The evaluated reputation scores is named as first-hand reputation scores. Specifically, in the data transmission stage of T_t , node N_i (N_iN) records the number of data packets sent to its next hop node N_j as $S_{ij}(t)$, and the number of data packets forwarded by N_j as $f_{ij}(t)$. Thus, the number of data packets lost in the transmission from N_i to N_j is $m_{ij}(t) = S_{ij}(t) - f_{ij}(t)$. Based on the

discussion of the previous subsection, we can estimate the normal packet loss rate between N_i and N_j as $p_{i,j}(t)$. Since each data packet is transmitted to N_j independently, the data transmission from N_i to N_j can be regarded as a sequence of independent repeated trials. It means, if N_i sends l data packets to N_j , the probability of k ($0 \leq k \leq l$) out of l packets lost during the transmission, denoted by $P_{i,j}(X = k)$, follows a binomial distribution, i.e.

$$P_{i,j}(X = k) = \binom{l}{k} (p_{i,j}(t))^k (1 - p_{i,j}(t))^{l-k}$$

We consider the forwarding behavior evaluation for N_j during an evaluation period T_i as a sampling test. If N_j behaves normally during data forwarding, $m_{i,j}(t)$ should slightly fluctuate around the estimated number of normal lost data packets $p_{i,j}(t) \cdot S_{i,j}(t)$. However, when $m_{i,j}(t) > p_{i,j}(t) \cdot S_{i,j}(t)$, with the increase of $m_{i,j}(t)$, the probability of N_j misbehaving in data forwarding increases. In order to evaluate $m_{i,j}(t)$, we introduce a detection threshold $i,j(t)$ ($S_{i,j}(t) \cdot p_{i,j}(t) < i,j(t) < S_{i,j}(t)$, $i,j(t)N+$) and define the reputation evaluation function of N_i to N_j as follows.

$$r^1_{i,j}(t) = \begin{cases} +\delta, & \text{if } m_{i,j}(t) \leq p_{i,j}(t) \cdot S_{i,j}(t) \\ -\delta, & \text{if } p_{i,j}(t) \cdot S_{i,j}(t) < m_{i,j}(t) \leq \xi_{i,j}(t) \\ -\lambda & \text{if } m_{i,j}(t) > \xi_{i,j}(t) \end{cases}$$

where λ is a punishment factor and δ is an adjustment factor. We set and explain the function as follows.

- If $m_{i,j}(t) \leq p_{i,j}(t) \cdot S_{i,j}(t)$, the sampling test is acceptable, which means the transmission between N_i and N_j is successful. Thus, N_i rewards a positive δ to N_j .
- If $p_{i,j}(t) \cdot S_{i,j}(t) < m_{i,j}(t) \leq \xi_{i,j}(t)$ we consider it is a normal fluctuation of $p^m_{i,j}$ around $p_{i,j}$, and rate to N_j to neutralize the reputation evaluation.
- When $m_{i,j}(t) > \xi_{i,j}(t)$ we consider there is a high probability for N_j to misbehave in the data forwarding. If it happens, N_i rates a punishment $-$ to N_j .

If N_j is a normal node, $m_{i,j}(t)$ will slightly fluctuate around $p_{i,j}(t) \cdot S_{i,j}(t)$. The proposed reputation evaluation function should make the reputation value of N_j stable or increased after a number of evaluation periods. On the other hand, if N_j misbehaves in



data forwarding, $m_{ij}(t)$ may be larger than $p_{ij}(t) \cdot S_{ij}(t)$ with a high probability. The proposed function should decrease the reputation value of N_j sharply after a number of evaluation periods.

3.3 Reputation Propagation

In order to share the monitored forwarding behavior information and hence to improve the attack detection accuracy, N_i propagates the first-hand reputation scores, such as $r_{ij}^1(t)$, to their neighbors during each T_t . The received reputation scores from the neighboring nodes are called as second-hand reputation scores, which reflect the evaluation of the neighboring nodes on their next hop nodes. However, the reputation propagation causes CRS-A vulnerable to collaborative promotion/demotion attacks, which means neighboring malicious nodes can collaborate with each other to mutually promote their reputation scores. To mitigate the impact of the potentially partial reputation scores, we determine the second-hand reputation scores as follows.

Denote the set of N_i 's neighboring sensor nodes as NC_i , and the number of nodes in NC_i as $|NC_i|$. We further divide the nodes of NC_i into two subsets, $NC_{i,g}$ and $NC_{i,b}$, based on their long-term reputation values in N_i . Let N_s be a node of NC_i . We put N_s into the honest neighbor set $NC_{i,g}$,

$$R_{i,s} > \frac{\sum_{x \in NC_i} R_{i,x}}{|NC_i|}$$

Otherwise, N_s is allocated to the dishonest neighbor set $NC_{i,b}$. Since the long-term reputation values of malicious nodes may decrease after misbehaving in a number of evaluation periods, these nodes are classified into the dishonest neighbor set and the weights of their propagating information are reduced by the penalty factor α . As a result, the negative impacts of mutual reputation promotions among neighboring malicious nodes can be significantly mitigated. To reduce the communication overhead of reputation propagation, the propagated reputation scores can be piggybacked to other data packets, such as the periodically exchanged neighbor information.

3.4 Reputation Integration

After reputation propagation, the first-hand and second-hand short-term reputation scores should be integrated to update the reputation table. Denote $R_{i,j}$ as the long-term reputation value of N_j in N_i 's reputation table, and R_m and R_s as the upper bound and lower bound of reputation value. We calculate the integrated reputation score as $R_{i,j}^1(t) = \sigma r_{i,j}^1(t) + (1 - \sigma) r_{i,j}^2(t)$, and update $R_{i,j}$ as the following equation.

$$R_{i,j} = \begin{cases} R_s, & \text{if } R_{i,j} + R_{li,j} \leq R_s \\ R_{i,j} + R_{li,j}, & \text{if } R_s < R_{i,j} + R_{li,j} < R_m \\ R_m, & \text{otherwise} \end{cases}$$

Here, σ is the weight factor of the first-hand information and $\sigma > 0.5$. R_m and R_s are system parameters that can be chosen based on the system requirements.

3.5 Malicious node identification

In each T_i , sensor nodes can evaluate the forwarding behaviors of their next hop sensor nodes and update their reputation table with the above three procedures. After a number of evaluation periods, the reputation values of malicious nodes are significantly reduced in the reputation tables of their neighboring nodes. To identify the malicious nodes, sensor nodes send their reputation tables to the sink for identification after a fixed time. When the average reputation value in N_j 's neighbors is below R_a , N_j is identified as a malicious node. Here, R_a is an alarm reputation value that can be predefined according to system requirements. If N_j is identified as a malicious node, the network operator can perform a security check or software reset for these nodes. However, since malicious nodes can mutually promote their reputation values or collaboratively degrade the reputation values of normal nodes, the average reputation value should be adjusted against the promotion and demotion attacks.

4. CRS with Adaptive Detection threshold



The detection accuracy of CRS-A is significantly impacted by the misbehaving detection threshold for reputation evaluation. In this section, we aim to determine the optimal evaluation threshold for each pair of neighboring nodes along the data forwarding path to optimize the detection accuracy of CRS-A. According to the attack model, malicious nodes can launch attacks with different probabilities, which indicate the detection threshold should be different for each communication link. Meanwhile, due to the nature of dynamic routing and time-varied channel condition in WSNs, the detection threshold should be adaptive to the time-varied data traffic and normal packet loss rate of the link. Without loss of generality, we focus on determining the optimal threshold for the transmission from N_i to N_j during the period T_t , in the following analysis.

Since CRS-A is proposed to detect selective forwarding attacks and identify malicious nodes, we first identify some performance metrics to evaluate CRS-A before optimizing them. If $\xi_{ij}(t)$ is set as a large value, the forwarding misbehavior of N_j will be regarded as a normal fluctuation, without being punished with . It means the attacks launched by N_j are not detected by the detection of CRS-A. On the other hand, if $\xi_{ij}(t)$ is set as a small value close to $S_{ij}(t) \cdot p_{ij}(t)$, the normal fluctuation of $m_{ij}(t)$ will be detected as a misbehavior, when N_j acts normally in data forwarding. It leads to a normal sensor node has a large probability to be falsely identified as a compromised node by the detection of CRS-A. Therefore, there exists a trade-off in determining the value of $\xi_{ij}(t)$ to optimize the detection accuracy for selective forwarding attacks.

Here we introduce two metrics, missed detection probability and false detection probability. The Missed Detection Probability is the probability that a malicious forwarding behavior is detected as a normal behavior, while the False Detection Probability refers to the probability that a normal forwarding behavior is detected as a malicious behavior. If we use X to denote the data packets lost in the transmission from N_i to N_j , and Y to denote the data packets maliciously dropped by N_j , the missed detection probability $\eta_{ij}(t)$ is

$$\eta_{ij}(t) = P\{X+Y \leq \xi_{ij}(t) / j \text{ misbehaved in } T_t\}$$



and the false detection probability $\mu_{i,j}(t)$ is

$$\mu_{i,j}(t) = P\{X+Y \leq \xi_{i,j}(t) / j \text{ behaved well in } T_t\}$$

Since both X and Y are discrete random variables, the probability mass function (PMF) of X and Y should be determined for calculating $\eta_{i,j}(t)$ and $\mu_{i,j}(t)$. X is defined as the number of normally lost data packets during the transmission. If the number of data packets sent by N_i during T_t is $S_{i,j}(t)$, the false detection probability $\mu_{i,j}(t)$ is the CDF of X .

However, due to $\eta_{i,j}(t)$ depending on the variable Y , we should determine the PMF of Y and $X + Y$. According to the attack model, each sensor nodes has a probability PM to be compromised by the adversary. It means $P\{Y = 0\} = 1 - PM$ and $P\{Y = Y'\} = PM$, where Y' is a discrete random variable denoting the number of maliciously dropped packets by N_j when N_j is a malicious node.

According to the attack model, when a malicious node successfully receives a data packet, it decides to maliciously drop the packet with a probability, which is called attack probability. We denote the attack probability of N_j as p_j . Since the number of data packets sent by N_i during the evaluation t are $S_{i,j}(t)$, the PMF of Y' should be a binomial function with the number of experiments as $A_i(t) = S_{i,j}(t)$. Obviously, $A_i(t)$ is a random variable depending on X , so we first calculate the conditional probability when $A_i(t)$ is fixed as a , ($0 \leq a \leq S_{i,j}(t)$, $0 \leq k \leq a$) as

$$P\{Y' = k | A_i(t) = a\} = \binom{a}{k} p_j^k (1-p_j)^{a-k}$$

And the PMF of Y' is

$$P\{Y' = k\} = \sum_{a=0}^{S_{i,j}(t)} P\{Y' = k | A_i(t) = a\} P\{A_i(t) = a\}$$

we can use the PMF of Y' to determine the PMF of Y as



$$P\{Y=k\} = \begin{cases} (1-PM) + PM.P\{Y'=0\}, & \text{if } k=0 \\ PM.P\{Y'=k\}, & \text{if } 1 \leq k \leq S_{i,j}(t) \end{cases}$$

If N_i sends $S_{i,j}(t)$ data packets to N_j during the evaluation period T_t and the detection threshold is $\xi_{i,j}(t)$ ($S_{i,j}(t) \cdot p_{i,j}(t) < \xi_{i,j}(t) < S_{i,j}(t)$), the missed detection probability for evaluating N_j is

$$\eta_{i,j}(t) = \sum_{k=1}^{S_{i,j}(t)} [P\{X \leq \xi_{i,j}(t) - k\} \cdot P\{Y=k\}, \text{ if } k=0]$$

$$P_M - P_M \cdot (1 - p_j)^{S_{i,j}(t)}$$

where $P\{X \leq k\}$ is the CDF of X .

Based on the PMF of X and Y , we further calculate $\eta_{i,j}$ as follows.

$$\eta_i = P\{\{X+Y \leq \xi_{i,j}(t)\} \cap \{Y > 0\}\}$$

$$P\{Y > 0\}$$

$$= \frac{P\{\{X+Y \leq \xi_{i,j}(t)\} \cap \{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}}{P\{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}}$$

$$P\{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}$$

$$= \frac{\sum_{k=1}^{S_{i,j}(t)} P\{\{X+Y \leq \xi_{i,j}(t)\} \cap \{Y=k\}\}}{P\{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}}$$

$$P\{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}$$

$$= \frac{\sum_{k=1}^{S_{i,j}(t)} [P\{X+Y \leq \xi_{i,j}(t) | Y=k\} \cdot P\{Y=k\}]}{P\{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}}$$

$$P\{\sum_{k=1}^{S_{i,j}(t)} \{Y=k\}\}$$

The missed detection probability $\eta_{i,j}(t)$ depends on the attack probability of N_j (i.e., p_j). Generally, the attack probabilities of malicious nodes are various and not known

by the system in advance. However, we can use the historical data to estimate p_j for each malicious node N_j . Specifically, in each T_t , N_i can estimate p_j

$$p_j = \frac{\left[\sum_{w=0}^t [m_{i,j}(w) - S_{i,j}(w) \cdot (1 - p_{i,j}(w))] \right]}{\sum_{w=0}^t [S_{i,j}(w) \cdot (1 - p_{i,j}(w))]}$$

Where $S_{i,j}(w) \cdot (1 - p_{i,j}(w))$ is the expected number of forwarded data packets at time period w , while $m_{i,j}(w) - S_{i,j}(w) \cdot (1 - p_{i,j}(w))$ is deviation between the actual number of forwarded data packets and the expected number of forwarded data packets at time period w . The probability that node j attacks (or maliciously drops) in data forwarding. When p_j is small or equal to 0, we consider N_j behaves well during the past data forwarding. The false detection probability μ_j should be minimized for CRS-A. As p_j keeps increasing, N_j has an increasing probability to be an attack. It indicates that the missed detection probability $\eta_{i,j}(t)$ should be emphasized to optimize the performance of CRS-A. Meanwhile, both of the missed detection probability $\eta_{i,j}(t)$ and false detection probability μ_j depend on $\xi_{i,j}(t)$. When $\xi_{i,j}(t)$ increases, $\eta_{i,j}(t)$ increases and μ_j decreases. And if $\xi_{i,j}(t)$ decreases, the situation reverses. It means $\eta_{i,j}(t)$ and μ_j are two contradictory optimization objectives. In order to find a trade-off between them, we can integrate $\eta_{i,j}(t)$ and μ_j as a single objective function v_j by weighting them with p_j and $1 - p_j$, respectively. The objective function is defined as $v_j = p_j \cdot \eta_{i,j}(t) + (1 - p_j) \cdot \mu_j$. Therefore, for each transmission from N_i to N_j in T_t , the optimal threshold determination problem can be formulated as calculating $\xi_{i,j}(t)$ to

$$(PP) \text{ minimize } v_j = p_j \cdot \eta_{i,j}(t) + (1 - p_j) \cdot \mu_{i,j}(t)$$

It is obvious that (PP) has only one optimization variable and a closed-form objective function. $\xi_{i,j}(t)$ is discrete, the objective function is non-differentiable with respect to $\xi_{i,j}(t)$, which indicates the hardness of deriving a closed-form optimal solution for (PP). However, due to the constraint that $\xi_{i,j}(t)$ should be an integer between $p_{i,j}(t) \cdot S_i(t)$ and $S_i(t)$, we can adopt a brute-force algorithm to calculate all the possible values for



determining the optimal one. Since $S_i(t)$ is the only input variable of (PP) which impacts the time complexity of finding a solution, the brute-force algorithm can guarantee the time complexity is $O(S_i(t))$, i.e., $O(n)$.

5. CSR-A with attack tolerant Data Forwarding

As a trust evaluation technique independent of route decision, CRS-A can be applied with any data forwarding protocol for WSNs. However, due to the negative impacts of selective forwarding attacks on data forwarding, data delivery ratio is a key performance metric for evaluating a defense technique, besides the detection accuracy for attacks and malicious nodes. We first develop a distributed and attack-tolerant data forwarding scheme to collaborate with CRS-A to improve the data delivery ratio of the network. Then, we summarize the main idea and procedures of CRS-A with attack-tolerant data forwarding into an algorithm.

For a distributed data forwarding scheme, the key challenge is to decide which sensor node should be chosen in the forwarding path to optimize the network performance, based on the local knowledge. In this, we consider data delivery ratio as the primary metric of network performance. Although we can detect the malicious nodes by CRS-A, it is unreasonable to isolate all the malicious nodes from the data forwarding path. We can illustrate it with the following Figure. N_a and N_b are two routing candidates of N_s , and N_a is identified as a malicious node by N_s . During T_t , N_s estimates the normal loss rate of each link as $p_{s,a}(t) = 10\%$ and $p_{s,b}(t) = 50\%$. The attack probabilities of N_a and N_b are $p_a = 20\%$ and $p_b = 0$, respectively. In this case, N_s have 6 data packets to forward. If N_s choose N_b as the next hop, the expected number of data packets that are successfully forwarded by N_b is 3. Contrastively, the expected number of data packets forwarded by N_a should be 5, even if its reputation in N_s is low and it has an attack probability 20% according to the historical records.

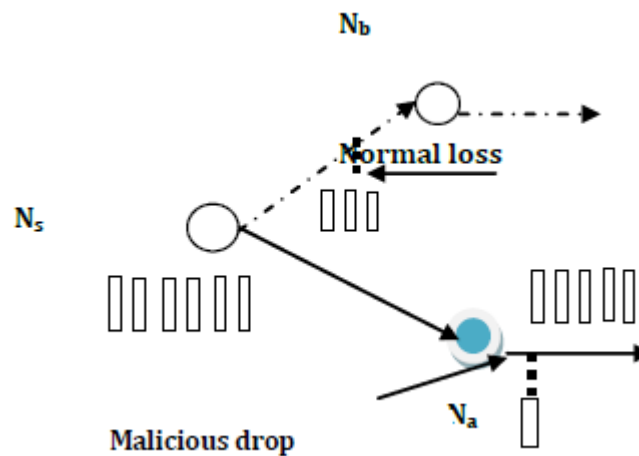


Figure 1: Example of Dynamic Routing

To select a better forwarding node to improve the data delivery ratio, we introduce the expected data forwarding ratio (DFR), which is defined as the ratio between the expected number of forwarded data packets and the total number of sent data packets. In each evaluation period T_t , N_i chooses the node with the highest DFR from its forwarding candidate set as the next hop. The forwarding candidate set of N_i is the set of its neighboring nodes that are geographically closer to the sink than N_i . Specifically, the forwarding decision can be formulated as follows. For each N_i , given the number of data packets that N_i transmits in T_t as $S_i(t)$, if choosing N_j as the data forwarding node, the expected number of lost data packets should be $L_j(t) = S_i(t) \cdot p_{ij}(t) + [S_i(t) - S_i(t) \cdot p_{ij}(t)] \cdot p_j(t)$. And, the DFR of N_j is

$$DFR_j(t) = (S_i(t)L_j(t))/S_i(t) = 1 - p_{ij}(t) - p_j(t) + p_{ij}(t) \cdot p_j(t)$$

Description: Updating the reputation of sensor nodes and data forwarding during T_t ($T_t T$).



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1 Phase I Normal Loss Estimation;
2 for each  $N_i \in N$  do
3 Estimate the normal packet loss rate  $p_{ij}(t)$  between  $N_i$  and
  each  $N_j$  in  $N_i$ 's neighbor set
4 end
5 Phase II Data Transmission and Monitoring;
6 for each  $N_i \in N$  do
7 Choosing  $N_j$  from  $RC_i$  as the next hop and use  $N_j$  to forward
  its data;
8 Record the number of sent data packets  $S_{ij}(t)$  and the
  number of data packets  $m_{ij}(t)$  forwarded by  $N_j$ ;
9 end
10 Phase III Reputation Evaluation and Updating;
11 for each  $N_i \in N$  do
12 Calculate the attack probability  $p_j$  of  $N_j$ ;
13 Determine the optimal detection threshold  $\xi_{ij}(t)$  by
  solving the problem (PP);
14 Evaluate the first-hand reputation score  $r^1_{ij}(t)$ 
15 Propagate  $r^1_{ij}(t)$  to its neighboring nodes;
16 if receive propagated reputation scores then
17 Calculate the second-hand reputation score  $r^2_{ij}(t)$ 
18 end
19 Calculate the integrated reputation score  $R^l_{ij}(t)$  with  $r^1_{ij}(t)$ 
  and  $r^2_{ij}(t)$  and update  $R_{ij}$ 
20 end
```



According to Algorithm 1, when a malicious node N_j is selected into the routing path by N_i , the evaluation threshold is determined by p_{ij} and p_j to evaluate its forwarding



behavior in the current evaluation period. If N_j misbehaves in this period with a probability p'_j that is higher than p_j , i.e., $p'_j > p_j$, the number of lost data packets will be larger than the evaluation threshold and it will be punished with a negative reputation score. Only if N_j adopts a lower attack probability, it could avoid a reputation punishment. For the irrational malicious nodes increasing the attack probability without considering the punishment, they are removed by the security check soon. Meanwhile, rational malicious nodes can be stimulated to behave better to achieve an improved data delivery ratio.

We consider the overhead of maintaining CRS-A, in terms of its storage overhead and communication overhead. In CRS-A, each node maintains a reputation table to record the reputation values of its neighboring nodes, which produces the storage overhead for sensor nodes. If the range of reputation value is set as $[0, 255]$, each reputation value only takes 8 bits and the total storage overhead of N_i for maintaining the CRS-A is $8 \cdot |NC_i|$ bits, where NC_i is the neighbor set of N_i . The communication overhead is mainly produced by channel estimation and reputation propagation. Let B be the number of bits in a PROBE packet that sensor nodes broadcast to their neighboring nodes for channel estimation. The overhead for channel estimation is B bits data broadcasting and $B \cdot |NC_i|$ bits data receiving for each node in an evaluation period. Similarly, each sensor node evaluates a reputation score for its data forwarding node, and propagates the score to its neighboring nodes in each evaluation period. Thus, the communication overhead of reputation propagation includes 8 bits data broadcasting and $8 \cdot |NC_i|$ bits data receiving. Since the PROBE packet and reputation score information are much smaller than the transmitted data packets of sensor nodes, it means CRS-A has a small communication overhead to be employed into WSNs.

6. Malicious Node Detection

To detect the malicious node we have proposed one method which uses a reactive routing protocol known as Ad hoc On demand Distance Vector (AODV) routing for analysis of the effect of the black hole attack when the destination sequence number is



changed via simulation. The proposed algorithm first detects those nodes, which may be malicious. Then the neighbor of the malicious node initiates a cooperative detection mechanism to detect the actual black hole node. In AODV routing, messages contain only the source and the destination addresses. It uses destination sequence numbers to specify the valid route. At first the sender broadcast the Route Request (RREQ) message to its neighbors. Each node that receives the broadcast, checks the destination to see if it is the intended recipient. If yes it sends a Route Reply (RREP) message back to the originator. RREP message contains the current sequence number of the destination node. The same process continues till the packets reach to destination or reach to an intermediate node, which has a fresh, enough routes to destination. Every node keeps track of its neighbor by maintaining two small size tables. One is sequence table (SnT) to keep the neighbor node's id and neighbor node's sequence number and other is the statustable (ST) to keep track of the node's status whether it is a safe node or a malicious one. Every node also maintains a neighbor list (N_List) and this list is updated periodically. When an intermediate node receives a RREP checks if the difference between the Dst_Seq present in the RREP message and the sequence no present in its table is greater than some predefined threshold value? if so then the intermediate node stops forwarding the message and mark the node as „M“ or malicious in the status table(ST) and send a notification message(NM) to source node along with the malicious node's id and neighbor list of the malicious node. The threshold value is the average difference of Dst_Seq in each time slot between the sequence number of RREP message and the one held in the table.

The source node has an additional table called Flag Table (FT). M1HN's after receiving the Further Detection message, broadcast a RREQ message by setting destination address to source node's address. If it receives a RREP message from the malicious node, it sends a Test packet (TP) to the source node via malicious node, and at the same time it sends a Acknowledgment Packet (AP) to source node(SN) though some other route. Then the source node waits for w time until it receives the entire test and acknowledgement packet. If, SN receives a TP, it updates the Flag Table (FT) by adding

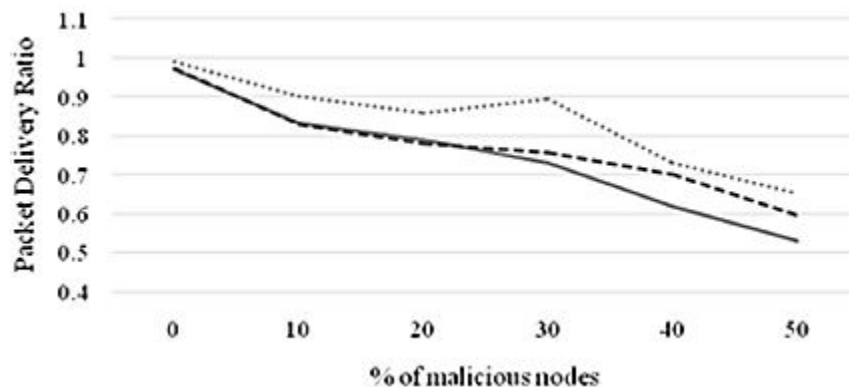
the source node id to the table and set the flag of the node as Y and if an AP is received set the flag as N and update the count field. If all the entries for the malicious node are N then source node updates the status table (ST) by adding the MN s id to the ST and making the status as B i.e. Black hole.

To accurately distinguish selective forwarding attacks from the normal packet loss, CRS- A evaluates the forwarding behaviors by the deviation between the estimated normal packet loss and monitored packet loss. To improve the detection accuracy of CRS-A, we have further derived the optimal evaluation threshold of CRS-A in a probabilistic way, which is adaptive to the time-varied channel condition and the attack probabilities of compromised nodes. In addition, a distributed and attack-tolerant data forwarding scheme is developed to collaborate with CRS-A for stimulating the cooperation of compromised nodes and improving the data delivery ratio.

7. Result and Discussion

7.1 Packet Delivery Ratio

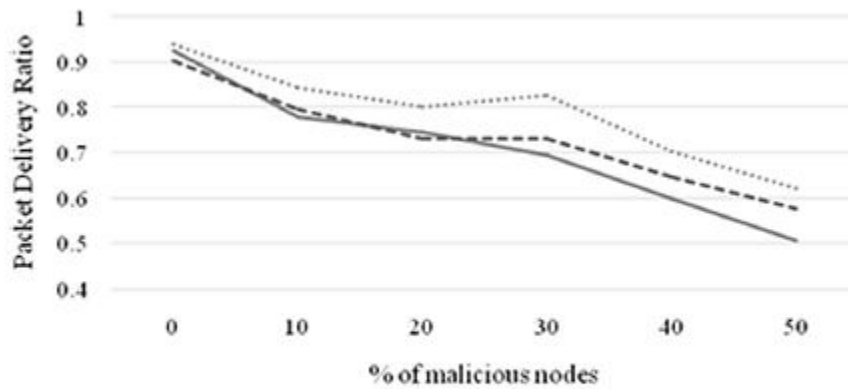
Following figures 2 gives the packet delivery ratio in 5m/s, figure 3 depicts the packet delivery ratio at 15m/s and figure 4 presents the packet delivery ratio at 25m/s. The packet delivery ratio is compared with the existing routing protocol like AODV and DSR against the percentage of malicious node.





— Intrusion Detection with AODV Protocol
- - - Intrusion Detection with DSR protocol
..... Intrusion Detection with Proposed Method

Figure 2: Packet Delivery ratio (5 m/s)



— Intrusion Detection with AODV Protocol
- - - Intrusion Detection with DSR protocol
..... Intrusion Detection with proposed method

Figure 3: Packet Delivery ratio (15 m/s)

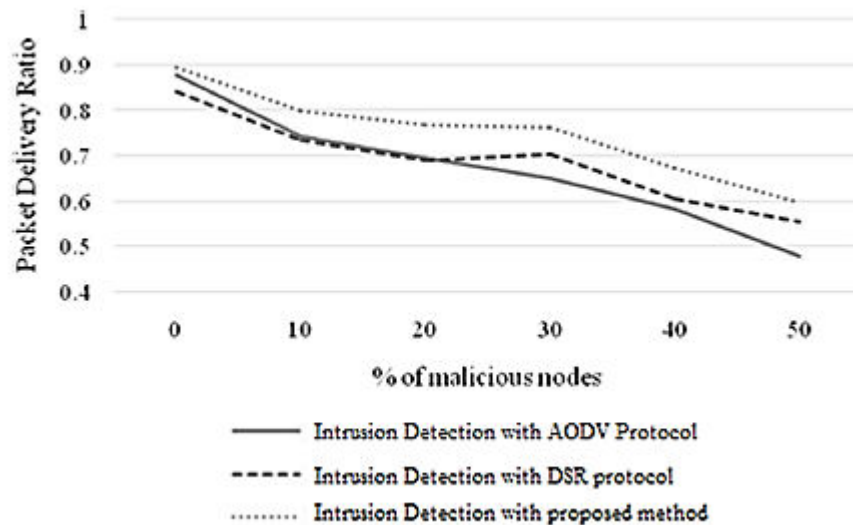


Figure 4: Packet Delivery ratio (25 m/s)

7.2 Routing Overhead

Following figures 5 gives the routing overhead in 5m/s, figure 6 depicts the routing overhead at 15m/s and figure 7 presents the routing overhead at 25m/s. The routing overhead is compared with the existing routing protocol like AODV and DSR against the percentage of malicious node.

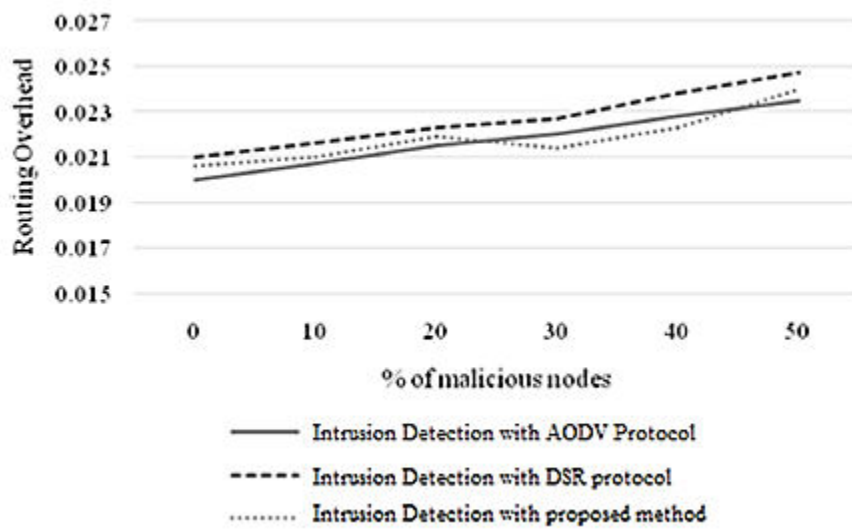


Figure 5: Routing overhead (5 m/s)

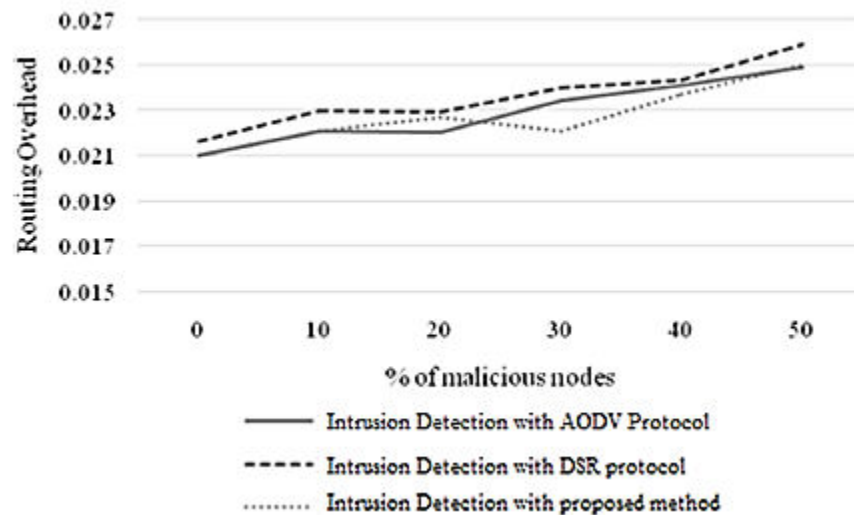


Figure 6: Routing Overhead (15m/s)

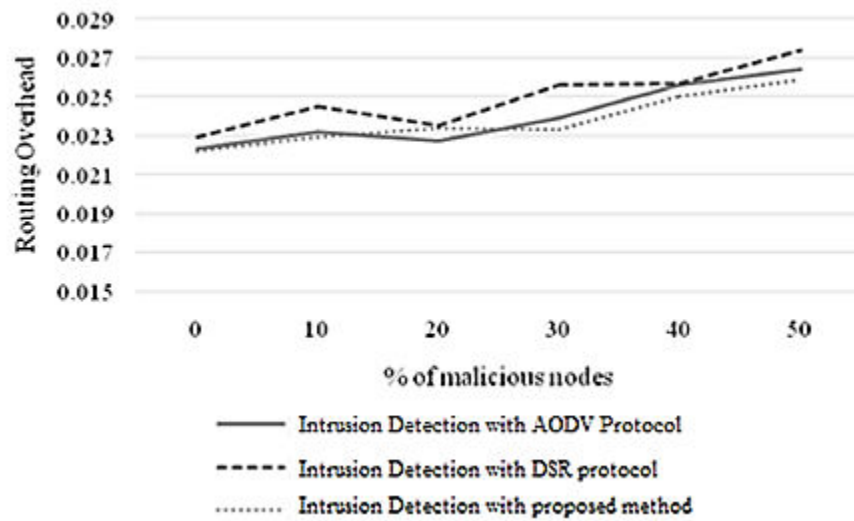


Figure 7: Routing Overhead (25m/s)

7.3 End to End Delay

Following figures 8 gives the end to end delay in 5m/s, figure 9 depicts the end to end delay at 15m/s and figure 10 presents the end to end delay at 25m/s. The routing overhead is compared with the existing routing protocol like AODV and DSR against the percentage of malicious node.

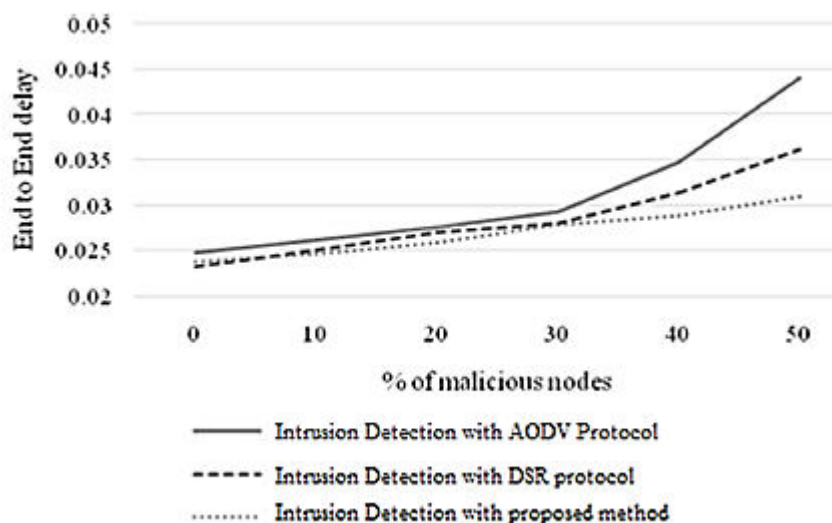


Figure 8: End to End Delay (5m/s)

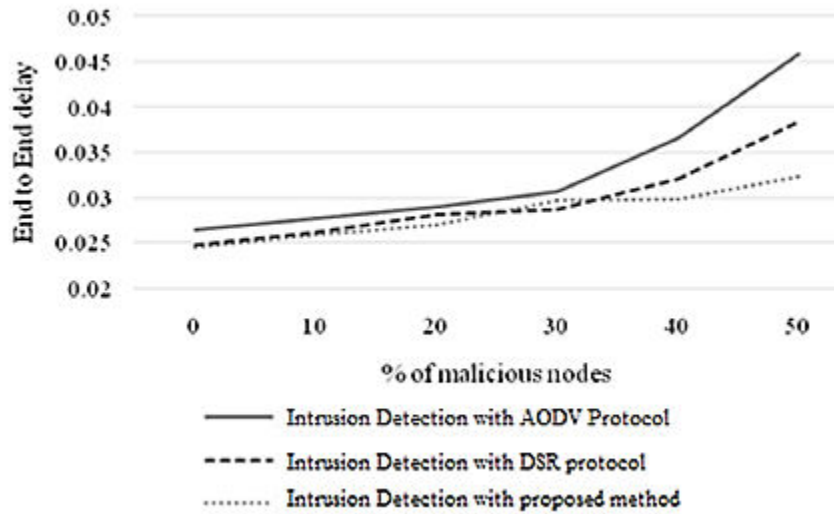


Figure 9: End to End Delay (15m/s)

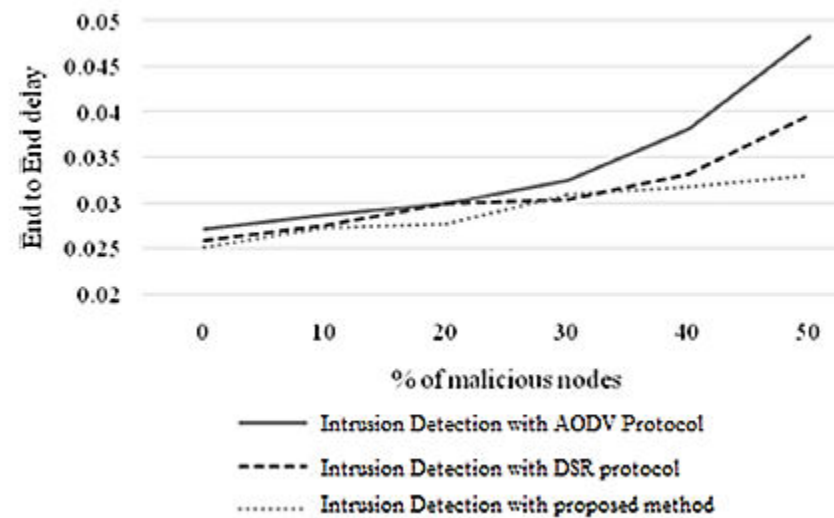


Figure 10: End to End delay (25m/s)

8. Conclusion



The open medium of WSN makes it vulnerable to various attacks. Various methods are incorporated to mitigate the malicious nodes from disrupting working of the network. Acknowledgment based routing solves ambiguous collisions, receiver collisions, and limited transmission power. Trust is an effective mechanism used for improving the security of wireless networks. In this paper, a new intrusion detection method for detecting selective forwarding attacks to predict the trust node in the network. This method is compared with other routing protocols like intrusion detection in AODV and DSR. This method is tested with various metrics like Packet Delivery Ratio (PDR), End to End Delay and Routing overhead against the number of malicious node in the network. From the above result obtained, it is clear that the proposed method performs better than the intrusion detection with routing protocols like AODV and DSR in terms of Packet delivery ratio, end to end delay and routing overhead.

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